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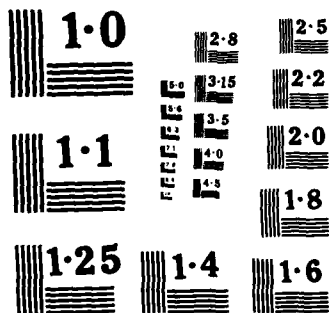
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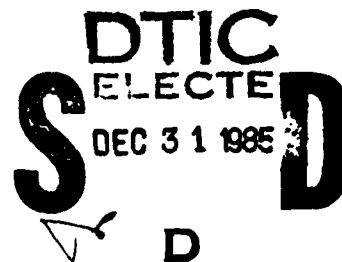
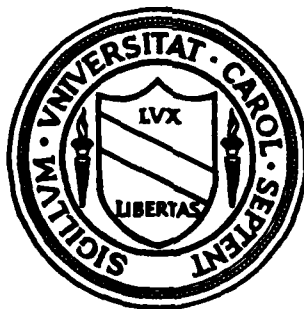
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by

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FLUCTUATIONS NEAR HOMOGENEOUS STATES OF CHEMICAL REACTIONS WITH DIFFUSION

PETER KOTELNEZ*, Universität Bremen

Abstract

Conditions are given under which a space-time jump Markov process describing the stochastic model of nonlinear chemical reactions with diffusion converges to the homogeneous state solution of the corresponding reaction-diffusion equation. The deviation is measured by a central limit theorem. This limit is a distribution valued Ornstein-Uhlenbeck process and can be represented as the mild solution of a certain stochastic partial differential equation.

REACTION-DIFFUSION EQUATION; STOCHASTIC MODEL OF NONLINEAR CHEMICAL REACTIONS WITH DIFFUSION; THERMODYNAMIC LIMIT; CENTRAL LIMIT THEOREM; HIGH DENSITY LIMIT; STOCHASTIC PARTIAL DIFFERENTIAL EQUATION

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1. Introduction

Mathematical models of chemical reactions have been described by Gardiner McNeil, Walls and Matheson [12], Haken [15], Nicolis and Prigogine [31], van Kampen [20], and Arnold [1]. For the deterministic theory of reaction-diffusion equations we refer to Smoller [33] and the references therein.

In [3] Arnold and Theodosopulu have constructed a space-time jump Markov process $X_{v,N}$ by dividing a finite interval I (one-dimensional reactor) into N cells, counting the number of particles in each cell and dividing this number by a proportionality factor v (the cell size of an unscaled model). This density changes in each cell due to reaction and diffusion (which couples neighbouring cells). The rates by which this density changes are derived from an underlying partial differential equation (PDE). Under a high density assumption ($\frac{N^2}{v} \rightarrow 0$) Arnold and Theodosopulu (loc.cit.) derived the law of large numbers (LLN) in $L_2(I)$, i.e. $X_{v,N} \rightarrow X$ in $L_2(I)$, where X is the solution of the PDE. In Kotelenez [22], [25] the corresponding central limit theorem (CLT) was proved under the assumption that the reaction is linear. In this linear case the density could be taken low, because the LLN was proved in distribution spaces (cf. also Kotelenez [26]). On the other hand, nonlinear operations like multiplication are not defined on distributions (cf. Schwartz [32]). Therefore it seems to be convenient - if not necessary - to prove for nonlinear chemical reactions with diffusion the LLN in a function space by making a high density assumption (as in Arnold and Theodosopulu [3]) and then derive the CLT in a distribution space. This, however, causes certain numerical difficulties (cf. our Remark 3.1) which do not show up if we assume that the deterministic limit X is spatially homogeneous (cf. (2.1) and (2.5)). This assumption allows us to derive the LLN (Theorem 3.1) in a function norm and the CLT (Theorem 3.3) in a distribution norm. The limit Y and the CLT is a generalized Ornstein-Uhlenbeck process (if Y_0 is Gaussian) and can be represented as the mild

solution of a certain stochastic partial differential equation (SPDE). We describe the optimal (smoothest) state spaces for Y . Our main tool is the calculus of stochastic evolution equations as developed in Kotelenetz ([21], [22], [24] - [27]) both for a fixed Hilbert state space and a nuclear Gel'fand triple (cf. (2.3)) as state space.

Apart from various Gaussian approximations to systems of (branching) Brownian motions (s. our references in Remark 2.2 - and also Kotelenetz [27]) we would like to mention the diffusion approximations to spatially distributed neurons given in Walsh [36] and Kallianpur and Wolpert [19], where the limit is also a generalized Ornstein-Uhlenbeck process, which can be interpreted as the solution of a linear SPDE (as in our case).

Let us briefly describe the contents. In Section 2 we introduce both the deterministic and the stochastic models on an n -dimensional unit cube. In the first part on the deterministic model we introduce the nuclear Gel'fand triple (2.3) and prove that the linear operators from our models can be "nicely" defined on the Hilbert distribution spaces in (2.3). In the second part on the stochastic model we derive some bounds on $X_{v,N}$ and its martingale part. In Section 3 we prove the LLN in sup-norm (Theorem 3.1) with a certain speed of convergence. Then we describe the limiting Gaussian martingale part for the normalized martingale parts of $X_{v,N}$, prove in several steps the CLT and describe the limit (Theorem 3.3).

2. The Models

Following Arnold and Theodosopulu [3] and Arnold [1] we first introduce the (local) deterministic model, then construct the corresponding (local) stochastic model, and finally compare the two models.

2.1 The (local) deterministic model

Set $S := \{q = (q_1, \dots, q_n) \in \mathbb{R}^n : 0 \leq q_i \leq 1, i = 1, \dots, n\}$. Let

$R(x) = b(x) - d(x) = \sum_{i=0}^m c_i x^i$ be a polynomial in $x \in \mathbb{R}$, where $c_0 \geq 0$, $c_m < 0$ and $b(x)$ and $d(x)$ are polynomials of degree $\leq m$ with nonnegative coefficients. Δ denotes the Laplacian and $D > 0$ a diffusion coefficient. Then the concentration of one reactant with reflection at the boundary is given by the solution of the following PDE:

$$(2.1) \quad \begin{cases} \frac{\partial}{\partial t} X(t, q) = D \Delta X(t, q) + R(X(t, q)) \\ \frac{\partial}{\partial q_i} X(t, q) = 0, \quad q_i \in \{0, 1\} \quad i = 1, \dots, n \\ X_0(q) \geq 0 \end{cases}$$

Let $\mathbb{H}_0 := L_2(S)$ be the Hilbert space of square integrable real valued functions on S equipped with the scalar product $\langle \varphi, \psi \rangle_0 := \int_S \varphi(q) \psi(q) dq$, $\varphi, \psi \in \mathbb{H}_0$. In what follows we shall denote by $D\Delta$ the closure of $D\Delta$ w.r.t. the reflecting boundary conditions of (2.1). $D\Delta$ is self-adjoint nonpositive on \mathbb{H}_0 and has a discrete spectrum. Let $\ell = (\ell_1, \dots, \ell_n)$ be a multiindex, where $\ell_i \in \mathbb{N} \cup \{0\}$, and set

$$\phi_{\ell_i} := \begin{cases} \sqrt{2} \cos \ell_i \pi(\cdot) & \ell_i \geq 1 \\ 1 & \ell_i = 0 \end{cases}$$

Then, the $\phi_{\ell} := \prod_{i=1}^n \phi_{\ell_i}$ are a complete orthonormal system (CONS) of eigenvectors of $D\Delta$ with eigenvalues $-D\mu_{\ell} := -D(\sum_{i=1}^n \ell_i^2 \pi^2)$. Consequently, the semigroup $T(t)$ generated by $D\Delta$ on \mathbb{H}_0 can be represented by

$$(2.2) \quad T(t)\varphi = \sum_{\ell} e^{-D\mu_{\ell}t} \phi_{\ell} \langle \varphi, \phi_{\ell} \rangle_0.$$

As in Kotelenetz ([26], [27]) we introduce the nuclear Gel'fand triple determined by $D\Delta$

$$(2.3) \quad \Phi \subset \mathbb{H}_{\alpha} \subset \mathbb{H}_0 = \mathbb{H}'_0 \subset \mathbb{H}_{-\alpha} \subset \Phi', \quad \alpha \geq 0.$$

In (2.3) we have $H_\alpha := \text{Dom}((I - D\Delta)^{\alpha/2})$, $\alpha \geq 0$ where I is the identity operator and "Dom" denotes "domain". H_α is a real separable Hilbert space if equipped with the scalar product $\langle \cdot, \cdot \rangle_\alpha := \langle (I - D\Delta)^{\alpha/2} \cdot, (I - D\Delta)^{\alpha/2} \cdot \rangle_0$ (for the definition of the α -th power of a positive self-adjoint operator - cf. Yosida [38]). H'_0 , the strong dual of H_0 , is identified with H_0 , $\Phi = \bigcap_{\alpha \geq 0} H_\alpha$ is a locally convex vector space whose topology is given by the set of norms $\{|\varphi|_\alpha := (\langle \varphi, \varphi \rangle_\alpha)^{1/2}, \varphi \in \Phi\}$, and Φ' is the strong dual of Φ . $H_{-\alpha}$ are those $\varphi' \in \Phi'$ which can be extended to continuous functionals on H_α , $\alpha \geq 0$. $H_{-\alpha}$ is a real separable Hilbert space with scalar product $\langle \cdot, \cdot \rangle_{-\alpha}$, where for φ , $\psi \in H_0$ $\langle \varphi, \psi \rangle_{-\alpha} = \langle (I - D\Delta)^{-\alpha/2} \varphi, (I - D\Delta)^{-\alpha/2} \psi \rangle_0$. Moreover, setting $\lambda_\ell := 1 + D\mu_\ell$, we obtain that

$$\phi_\ell^\alpha := \lambda_\ell^{-\alpha/2} \phi_\ell$$

is a CONS for H_α , $\alpha \in \mathbb{R}$. Hence

$$H_\alpha = \{\varphi' \in \Phi' : \sum_\ell (\varphi', \phi_\ell)^2 \lambda_\ell^\alpha < \infty\},$$

where (\cdot, \cdot) denotes the dual pairing. Thus, if we set

$$l_{2,\alpha} := \{(a_\ell) \in \mathbb{R}^\infty : \sum_\ell a_\ell^2 \lambda_\ell^\alpha < \infty\}$$

we see that (2.1) can be identified with a subset of \mathbb{R}^∞ , where H_α is isomorphic to $l_{2,\alpha}$, $\alpha \in \mathbb{R}$. Clearly, the imbeddings in (2.3) are continuous and dense.

Lemma 2.1

For any $\alpha, \gamma \in \mathbb{R}$ s.t. $\alpha > \gamma + \frac{n}{2}$ the imbedding

$$H_\alpha \rightarrow H_\gamma$$

is Hilbert-Schmidt.

Proof

$$\sum_{\ell} |\phi_{\ell}^{\alpha}|_{\gamma}^2 = \sum_{\ell} |\phi_{\ell}^{\alpha-\gamma}|_0^2 < \infty \text{ iff}$$

$$\int_1^{\infty} \dots \int_1^{\infty} (1+x_1^2 + \dots + x_n^2)^{-\alpha+\gamma} dx_1 \dots dx_n < \infty \text{ iff}$$

$$\int_1^{\infty} \dots \int_1^{\infty} (1+x_1 \dots + x_n)^{-2\alpha+2\gamma} dx_1 \dots dx_n < \infty \text{ iff}$$

$$2\alpha > n + 2\gamma.$$

□

Since $(I-D\Delta)^{\alpha/2}$ and $T(t)$ commute, $T(t)$ can be extended (resp. restricted) to a strongly continuous semigroup $T_{\alpha}(t)$ on H_{α} , $\alpha \in \mathbb{R}$, s.t. for all $\alpha \in \mathbb{R}$

$$(2.4) \quad \|T_{\alpha}(t)\|_{L(H_{\alpha})} = \|T(t)\|_{L(H_0)} \leq 1.$$

$\|\cdot\|_{L(H_{\alpha})}$ denotes the usual operator norm on H_{α} , and the inequality in (2.4) holds because $D\Delta$ is dissipative (cf. Davies [9]). Let us denote by $D\Delta_{\alpha}$ the generator of $T_{\alpha}(t)$ (which is the extension (resp. restriction) of $D\Delta$). As in Kotelenetz [26], Lemma 2.2, we obtain:

Lemma 2.2

For all $\alpha \in \mathbb{R}$, $\text{Dom}(D\Delta_{\alpha}) = H_{\alpha+2}$ and $T_{\alpha}(t)$ is analytic.

For the rest of the paper we shall assume that the system (2.1) starts in a spatially homogeneous state $X_0 = \rho_0 > 0$. This implies that the solution $X(t,r)$ of (2.1) is spatially homogeneous, i.e. $X(t,r) \equiv \rho(t)$ satisfies the ordinary differential equation

$$(2.5) \quad \frac{d}{dt} \rho(t) = R(\rho(t)), \quad \rho(0) = \rho_0 > 0$$

(cf. Arnold [1]). (2.5) has a unique positive bounded solution which

is strictly positive for all $t \geq 0$ (cf. Coddington and Levinson [5]).

2.2 The (local) stochastic model

We cover S with grid of N n -dimensional cubes (cells) of size h^n which are parallel to the axes. The cell corresponding to the grid point r^j is defined by

$$[r^j] := \{r \in S: r_i^j \leq r_i < r_i^j + h, i = 1, \dots, n\}, j = 1, \dots, N.$$

Let v be a parameter (which is explained in Remark 2.1) and denote by E_N the (countable) state space of elements $k = (k_{r^j})_{([r^j] \in S)}$, where $k_{r^j} \in \frac{N}{v} \cup \{0\}$. Set

$$H_{O,N} := \{\varphi \in H_O : \varphi \text{ constant on each } [r^j]\}.$$

Then

$$E_N \subset H_{O,N}$$

and

$$\pi_N := H_O \rightarrow H_{O,N} \text{ defined by}$$

$$\pi_N \varphi(r) := \varphi_N(r) := h^{-n} \int_{[r^j]} \varphi := h^{-n} \int_{[r^j]} \varphi(q) dq \text{ if } r \in [r^j]$$

is a projection from H_O onto $H_{O,N}$.

Now we define a Markov chain on E_N through the Q -matrix of its transition intensities:

$$(2.6) \quad \beta(k, m) := \begin{cases} vb(k_{r^j}) & m = k + e_{r^j} & =: m_{+0} \\ vd(k_{r^j}) & m = k - e_{r^j} & =: m_{-0} \\ vDh^{-2}k_{r^j} & m = k + e_{r^j \pm h_i} - e_{r^j} & =: m_{\pm i}, i = 1, \dots, n \\ & \text{for } 0 \leq r^j \pm h_i \leq 1 \\ - \sum_{i=0}^n \beta(k, m_{\pm i}) & m = k \\ 0 & \text{otherwise.} \end{cases}$$

Here $e_{rj} = (1_{[rj]})$, where $1_{[rj]}(q) = 1$ if $q \in [rj]$, $= 0$ otherwise, and $h_i = (0, \dots, 0, h, 0, \dots, 0)$ where all but the i -th coordinate are zero. Hence, we obtain the distributions $P(t, k)$ determined by $Q = (\beta(k, m)_{k, m \in \mathbb{E}_N})$ as the unique solution of Kolmogorov's backward equation (which is called in the application-oriented literature the "multivariate Master equation") (cf. Arnold [1]). The corresponding (canonical) cadlag Markov process will be denoted by

$$(2.7) \quad \begin{cases} X_{v, N} & \text{(generated by } Q) \\ X_{v, N}^{(0)} = X_{v, N, 0} \in \mathbb{E}_N & \text{a given initial distribution.} \end{cases}$$

In what follows we shall assume that the stochastic basis for $X_{v, N}$ $(\Omega, \mathcal{F}, \mathcal{F}_{v, N, t}, P)$ is complete with right continuous filtration.

Remark 2.1

We can view $X_{v, N}$ as the rescaled density Markov process of Arnold [1] and Arnold and Theododopulu [3] on a cube of volume $vN = V$ with cells of size v , where the number of particles is proportionel to v .

Remark 2.2

If $b(r) = br + c_0$, $d(r) = dr$, for some constants $b, d > 0$ then $X_{v, N}$ is a branching diffusion with immigration (c_0) on the grid. This case was investigated in Kotelenetz [25], [26], and the limit theorems therein corresponded to limit theorems for branching Brownian motions obtained by Holley and Stroock [16], Gorostiza [14] and, in the absence of branching ($b = d = c_0 = 0$) to Martin-Löf [30] and Itô [18] (cf. also Walsh [37]). For a diffusion approximation to branching Brownian motions - cf. Dawson [11].

In what follows we shall not explicitly write the parameter v , i.e., we shall write X_N instead of $X_{v,N}$ etc.

Extend $\varphi_N \in \mathbb{H}_{0,N}$ by reflection to

$$S_h := \{r \in \mathbb{R}^n : -h \leq r_i \leq 1+h, i = 1, \dots, n\}$$

and set

$$\nabla_{\pm i}^N \varphi_N := h^{-1} [\varphi_N(r \pm h_i) - \varphi_N(r)]$$

$$\Delta_N \varphi_N := - \sum_{i=1}^n \nabla_{-i}^N \nabla_i^N \varphi_N.$$

Remark 2.3

In view of our boundary condition we easily see that $D\Delta_N$ is selfadjoint and dissipative both as an operator on $\mathbb{H}_{0,N}$ and \mathbb{H}_0 (where on \mathbb{H}_0 Δ_N is defined by $\Delta_N \circ \pi_N$). If we set

$$\bar{\phi}_{\ell_i, N} := \frac{\phi_{\ell_i, N}}{|\phi_{\ell_i, N}|_0}$$

we see that

$$\{\bar{\phi}_{\ell, N} := \prod_{i=1}^n \bar{\phi}_{\ell_i, N}, \ell_i < h^{-1}\}$$

is a CONS of eigenvectors of $D\Delta_N$ for $\mathbb{H}_{0,N}$ with eigenvalues

$$D\mu_{\ell, N} := 2^n D N^2 \prod_{i=1}^n \{1 - \cos \ell_i h \pi\}.$$

The waiting time parameter for X_N is given by

$$\sigma_N(k) = v \sum_{r \in S} |R|(k_{rj}) + h^{-2} \sum_{i=1}^n 2Dk_{rj}$$

with $|R|(x) = b(x) + d(x)$, $x \in \mathbb{R}$. Hence, if

$$\Theta_N(k, m) := (\sigma_N(k))^{-1} \beta(k, m)$$

denotes the jump distribution function ($\beta(k,m)$ from (2.6)), then the infinitesimal generator for X_N is given by

$$(2.8) \quad (Af)(k) = \sigma_N(k) \int_{\mathbb{E}_N} [f(m) - f(k)] \theta_N(k, dm),$$

where $f: \mathbb{E}_N \rightarrow \mathbb{R}$ is bounded and measurable (Gihman and Skorohod [13]).

Let

$$\|\phi_N\| := \sup_{r \in S} |\phi_N(r)|$$

be the sup-norm on $\mathbb{H}_{0,N}$. If there is a finite constant $K(v,N)$ s.t.

$$(2.9) \quad \|X_{v,N}(0)\| \leq K(v,N) \quad \text{a.s.}$$

then by a lemma of Kurtz [28] (cf. Arnold and Theodosopulu [3] and Kotelenez [25])

$$(2.10) \quad \begin{cases} Z_N(t) := X_N(t) - X_N(0) - \int_0^t \sigma_N(X_N(s)) \int_{\mathbb{E}_N} (z - X_N(s)) \theta_N(X_N(s), dz) ds \\ = X_N(t) - X_N(0) - \int_0^t [D\Delta_N X_N(s) + R(X_N(s))] ds \end{cases}$$

is an $\mathbb{H}_{0,N}$ -valued square integrable cadlag martingale.

We shall assume (2.9) throughout the paper.

Hence, X_N satisfies formally the stochastic evolution equation

$$(2.11) \quad \begin{cases} dx_N(t) = [D\Delta_N X_N(t) + R(X_N(t))]dt + dz_N(t) \\ X_N(0) = X_{N,0} \end{cases},$$

and the difference $X_N(t) - X(t)$, where X is the solution of (2.1)/(2.5), satisfies

$$(2.12) \quad \left\{ \begin{aligned} X_N(t) - X(t) &= X_N(0) - X(0) + \int_0^t (DA_N + R'(X(s))(X_N(s) - X(s))) ds \\ &+ \int_0^t (X_N(s) - X(s))^2 \tilde{R}(X_N(s), X(s)) ds \\ &+ Z_N(t) \end{aligned} \right.$$

$R'(x)$ is the derivative of $R(x)$, $\tilde{R}(y, x)$ is a polynomial in y and x of degree $\leq m-2$, and $R'(X(s))$ and $(X_N(s) - X(s))^2$ are interpreted as multiplication operators.

Note that both $DA_N + R'(X(s))$ and $DA + R'(X(s))$ are quasi-generators of evolution operators $U_N(t, s)$ and $U(t, s)$ on $H_{0, N}$ and H_0 , respectively. (For the definition of evolution operators $V(t, s)$, i.e., strongly continuous two-parameter semigroups - cf. Curtain and Pritchard [7] and Tanabe [34], where $V(t, s)$ is called fundamental solution - any strongly continuous one-parameter semigroup is, of course, also an evolution operator.) Consequently, by variation of constants, (2.12) yields

$$(2.13) \quad \left\{ \begin{aligned} X_N(t) - X(t) &= U_N(t, 0)(X_N(0) - X(0)) + \int_0^t U_N(t, s) dZ_N(s) \\ &+ \int_0^t U_N(t, s)(X_N(s) - X(s))^2 \tilde{R}(X_N(s), X(s)) ds. \end{aligned} \right.$$

In order to give a meaning to the stochastic convolution integral in (2.13) we recall from Kotelenetz [21], [24]:

Definition 2.1

Let H be a separable Hilbert-space with Hilbert space norm $|\cdot|_H$ and $V(t, s)$ an evolution operator on H , $0 \leq s \leq t < \infty$. $V(t, s)$ is of contraction-type or, equivalently, $V(t, s) \in G(1, \beta_t)$ if for all $\hat{t} > 0$ there is a finite constant $\beta_{\hat{t}} \geq 0$ s.t.

$$(2.14) \quad |V(t, s)|_{L(H)} \leq e^{\beta_{\hat{t}}(t-s)}$$

for all $0 \leq s \leq t \leq \hat{t}$.

Remark 2.4

Let M be an H -valued locally square integrable cadlag martingale and $V(t,s) \in G(1, \beta_{\hat{t}})$ on H . Then, from Kotelenetz [21], we have

- (i) If M is cadlag, then $\int_0^\cdot V(\cdot, s) dM(s)$ has a cadlag version; if M is continuous, then $\int_0^\cdot V(\cdot, s) dM(s)$ has a continuous version.

A partial result from Kotelenetz [24] is the following:

- (ii) If $V(t,s)$ has a quasi-generator $A(t)$ and $\text{Dom}(A(t))$ is independent of t then for all $\hat{t} > 0$ there is a finite constant $c = c(\hat{t}, \beta)$ depending only on the scalar product $\langle \cdot, \cdot \rangle_H$, \hat{t} and β s.t. for all $t \leq \hat{t}$

$$(2.15) \quad E \sup_{0 \leq s \leq t} \left| \int_0^s V(s, u) dM(u) \right|_H^2 \leq c e^{4\beta \hat{t}} E |M(t)|_H^2.$$

For more general properties and inequalities for stochastic convolution integrals cf. Kotelenetz [24], [27].

Since $X(t)$ is constant in the space variable (cf. (2.5)) we obtain

$$(2.16) \quad \begin{cases} U_N(t,s) = T_N(t-s) \exp\left(\int_s^t R'(X(u)) du\right) \\ U_\alpha(t,s) = T_\alpha(t-s) \exp\left(\int_s^t R'(X(u)) du\right), \quad \alpha \in \mathbb{R}, \end{cases}$$

where the last equation means that $U(t,s)$ is extendible (resp. restrictable) to the H_α . Let us denote by $(H_{\alpha, N}, \langle \cdot, \cdot \rangle_\alpha)$ $H_{0, N}$ equipped with the Hilbert norm $\langle \cdot, \cdot \rangle_\alpha$, $\alpha \in \mathbb{R}$. Set

$$\beta := \sup_{0 \leq t < \infty} R'(X(t))$$

and note that $\beta < \infty$ by our assumption on $R(x)$ (cf. (2.5)).

Lemma 2.3

For all $\alpha \in \mathbb{R}$

$$U_{\alpha}(t,s) \in G(1,\beta) \text{ on } H_{\alpha}$$

$$U_N(t,s) \in G(1,\beta) \text{ on } H_{\alpha,N}.$$

Proof

(i) The statement for $U_{\alpha}(t,s)$ follows from (2.4) and (2.16).

(ii) Let $x \in H_{0,N}$. Then

$$\begin{aligned} |U_N(t,s)x|_{\alpha}^2 &= \sum_{\ell} \langle U_N(t,s)x, \phi_{\ell} \rangle_0^2 \lambda_{\ell}^{\alpha} \\ &= \sum_{\ell} \langle x, U_N(t,s)\phi_{\ell,N} \rangle_0^2 \lambda_{\ell}^{\alpha} \\ &\leq \sum_{\ell} e^{2\beta(t-s)} \langle x, T_N(t-s)\phi_{\ell,N} \rangle_0^2 \lambda_{\ell}^{\alpha} \\ &\quad \text{by (2.16)} \\ &\leq e^{2\beta(t-s)} |x|_{-\alpha}^2, \end{aligned}$$

since $T_N(t)$ is a contraction and $\phi_{\ell,N}$ is an eigenvector of $T_N(t)$.

□

The previous considerations show that the limit behaviour of $X_N(t) - X(t)$ essentially depends on the limit behaviour of $Z_N(t)$, $U_N(t,s)$ and the last term in (2.13).

We shall first give an estimate on the variance of $Z_N(t)$. To this end we define an operator on H_0 by

$$(2.17) \quad F_N(\varphi) := D \sum_{i=1}^n \nabla_{-i}^N \varphi \nabla_i^N + \nabla_i^N \varphi \nabla_{-i}^N + |R|(\varphi)$$

where $\varphi \in H_0$ and $|R|(\varphi)$ act as multiplication operators. As in the linear case of Kotelenetz [25] we obtain

Lemma 2.4

For arbitrary $\varphi \in H_0$

$$(2.18) \quad E \langle Z_N(t), \varphi \rangle_0^2 = \frac{1}{vN} \int_0^t E \langle F_N(X_N(s)) \varphi_N, \varphi_N \rangle_0 ds.$$

We need estimates on $X_N(t, r)$ which satisfies by variation of constants

$$(2.19) \quad X_N(t) = T_N(t)X_N(0) + \int_0^t T_N(t-s)dZ_N(s) + \int_0^t T_N(t-s)R(X_N(s))ds.$$

Set $\bar{\rho} := \max \{R(x) : x \in \mathbb{R}_+\}$.

The definition of $R(x)$ implies $\bar{\rho} < \infty$.

Lemma 2.5

For any $t > 0$

$$(2.20) \quad \sup_{0 \leq s \leq t} ||| E X_N(s) ||| \leq t\bar{\rho} + ||| E X_N(0) ||| ,$$

$$(2.21) \quad ||| X_N(t) ||| \leq t\bar{\rho} + ||| X_N(0) ||| + \sqrt{N} \left| \int_0^t T_N(t-s)dZ_N(s) \right|_0 .$$

Proof

- (i) From Kotelenetz [22], Lemma A.7 and Davies [9], Th. 7.16 we obtain that $T_N(t)$ is positivity-preserving on $H_{0,N}$, i.e., leaves the cone of nonnegative functions invariant, which implies (2.20).

(ii) We easily check that for any $\varphi_N \in H_{0,N}$

$$(2.22) \quad ||| \varphi_N ||| \leq \sqrt{N} |\varphi_N|_0,$$

whence we obtain (2.21) from (2.19).

□

3. Limit Theorems

Theorem 3.1 (LLN)

Assume (2.9) in addition to

(I) $v = N^p$, where $p > \frac{2\gamma+1+\frac{2}{n}}{1-2\gamma}$ and $\gamma \in [\frac{1}{4}, \frac{1}{2})$ arbitrary and fixed ;

(II) $E ||| (vN)^\gamma (X_N(0) - X(0)) ||| \rightarrow 0$, as $N \rightarrow \infty$.

Then for all $\hat{t} > 0$, $\delta > 0$

$$P\{ \sup_{0 \leq t \leq \hat{t}} ||| (vN)^\gamma (X_N(t) - X(t)) ||| > \delta \} \rightarrow 0; \text{ as } N \rightarrow \infty.$$

Proof

(i) (2.18), (2.20) and our assumptions imply the existence of a finite constant K s.t. for any $t \geq 0$

$$(3.1) \quad E |Z_N(t)|_0^2 \leq \frac{K(t+1)}{v} (c_0 + N^{2/n}),$$

where by (2.15) and Lemma 2.3 there is for any $\hat{t} \geq 0$ a finite constant $K(\beta, \hat{t})$ s.t. for $\gamma \in [\frac{1}{4}, \frac{1}{2})$ and

$$\eta_N(t) := (vN)^\gamma N^{1/2} \max \{ |\int_0^t U_N(t,s) dZ_N(s)|_0, |\int_0^t T_N(t-s) dZ_N(s)|_0 \}$$

$$\begin{aligned} E \sup_{0 \leq t \leq \hat{t}} \eta_N^2(t) &\leq K(\beta, \hat{t}) \frac{(vN)^{2\gamma}}{v} N(c_0 + N^{2/n}) \\ &= K(\beta, \hat{t}) N^{p(2\gamma-1) + 2/n + 1 + 2\gamma} \\ &\rightarrow 0, \text{ as } N \rightarrow \infty \text{ by the definition of } p. \end{aligned}$$

(ii) (2.16) implies that $U_N(t, s)$ is positivity preserving since $T_N(t)$ is positivity-preserving. Abbreviating

$$\zeta_N(t) := e^{\beta t} (vN)^\gamma ||| X_N(0) - X(0) ||| + \eta_N(t)$$

and

$$\psi_N(t) := e^{\beta \hat{t}} ||| \mathcal{R}(X_N(t), X(t)) ||| \cdot ||| X_N(t) - X(t) |||, \quad t \leq \hat{t}$$

the Gronwall-Bellmann lemma and (2.13) imply

$$(vN)^\gamma ||| X_N(t) - X(t) ||| \leq \zeta_N(t) + \int_0^t \zeta_N(s) \psi_N(s) \exp\left(\int_0^t \psi_N(u) du\right) ds, \quad t \leq \hat{t}.$$

Since by step (i) $\sup_{0 \leq t \leq \hat{t}} \psi_N(t)$ is stochastically bounded as $N \rightarrow \infty$ and

$\sup_{0 \leq t \leq \hat{t}} \zeta_N(t)$ tends to zero in mean square the proof is finished. □

Set

$$M_N := (vN)^{1/2} Z_N$$

and define for $\varphi \in \mathcal{H}_1$ the continuous analogue to (2.17):

$$(3.2) \quad F(\varphi) := -2D \sum_{i=1}^n \partial_i \varphi \partial_i + |R|(\varphi)$$

where again φ and $|R|(\varphi)$ act as multiplication operators. Denote for $\mu \in (0, 1)$ by

$$C^\mu([0, \infty); \mathcal{H})$$

the space of Hölder continuous H -valued functions with Hölder exponent μ , where H is some Hilbert space.

Lemma 3.1

There is a unique (in distribution) Φ' -valued Gaussian martingale M on some probability space $(\tilde{\Omega}, \tilde{F}, \tilde{F}_t, \tilde{P})$ with characteristic functional

$$(3.3) \quad \tilde{E} \exp(i(M(t), \varphi)) = \exp(-\frac{1}{2} \int_0^t \langle F(X(s)) \varphi, \varphi \rangle ds),$$

$\varphi \in \Phi$, where \tilde{E} is the mathematical expectation w.r.t. \tilde{P} . Moreover, for any $\alpha > \frac{n}{2} + 1$ and any $\mu \in (0, \frac{1}{2})$

$$M \in C^\mu([0, \infty), H_{-\alpha}) \quad \text{a.s.}$$

The proof of the existence and uniqueness is given in Itô [17] (cf. also Ustunel [35]), and the Hölder continuity follows from Kotelenetz [23].

Since $X(t)$ is spatially homogeneous and strictly positive we easily check that $F(X(t))$ as a positive self-adjoint operator on H_0 is just equal to $-2X(t)D\Delta + |R|(X(t))$, which has the same eigenfunctions ϕ_ℓ as $D\Delta$. Thus, the square root of $F(X(t))$ can be considered as an element $F_\alpha^{1/2}(X(t))$ from $L(H_\alpha, H_{\alpha-1})$ for all $\alpha \in \mathbb{R}$. If $\alpha > \frac{n}{2}$ then there is an $H_{-\alpha}$ valued Wiener process $W(t)$ on $H_{-\alpha}$ which is the cylindrical Brownian motion on H_0 (cf. Itô [17]). We may without loss of generality assume that $W(t)$ is also defined on $(\tilde{\Omega}, \tilde{F}, \tilde{F}_t, \tilde{P})$ from Lemma 3.1. Repeating now the proof of Lemma 2.4 in Kotelenetz [25] we obtain

Lemma 3.2

$$(3.4) \quad M \stackrel{D}{=} \int_0^\cdot F_{-\alpha+1}^{1/2}(X(s)) dW(s) \quad (\text{equal in distribution})$$

on $C^\mu([0, \infty); H_{-\alpha-1})$ for all $\alpha > \frac{n}{2}$, all $\mu \in (0, \frac{1}{2})$.

Let us denote by $D([0, \infty); H)$ the complete metric space of H -valued cadlag functions, where H is a separable Hilbert space (i.e. the Skorohod space - cf. Billingsley [4] and Kurtz [29]) and by " \Rightarrow " weak convergence.

Lemma 3.3

Under the assumptions of Theorem 3.1 for all $\alpha > \frac{n}{2} + 1$

$$M_N \Rightarrow M \text{ on } D([0, \infty); H_{-\alpha}),$$

where M is the Gaussian martingale given in Lemma 3.1.

Proof

- (i) The weak convergence of $M_N(t)$ to $M(t)$ for fixed t follows as in the linear case (cf. Kotelenetz [25], [26]).
- (ii) We shall estimate the "modules of continuity". Set for some (large) $K > 0$

$$\tau_N := \inf\{t \geq 0 : ||| X_N(t) ||| \geq K\}$$

and $\tilde{M}_N(t) := M_N(t \wedge \tau_N)$

where " \wedge " denotes "min". Then, abbreviating $F_{N,t} := \sigma(X_N(s), s \leq t)$ we obtain for $t \leq \hat{t}$, $s > 0$

$$\begin{aligned} & E \{ |M_N(t+s) - M_N(t)|_{-\alpha}^2 \mid F_{N,t} \} \\ (*) & \leq E \{ |\tilde{M}_N(t+s) - \tilde{M}_N(t)|_{-\alpha}^2 \mid F_{N,t} \} \\ & + E \{ 1_{\{\tau_N < \hat{t}+s\}} \mid F_{N,t} \}. \end{aligned}$$

Take the CONS $\{o_\ell^\alpha\}$ for H_α . By (2.18) (cf. Kurtz [28] and Kotelenetz [22], [25] for the step from the unconditional to the conditional expectation) the first term in the r.h.s. of (*) can be estimated from above by

$$(**) \begin{cases} \sum_{\ell} E \left\{ \int_t^{t+s} \langle \phi_{\ell}^{\alpha}, F_N(X_N(u \wedge \tau_N)) \phi_{\ell}^{\alpha} \rangle_0 du \mid F_{N,t} \right\} \\ \leq 2 \sum_{\ell} E \left\{ \int_t^{t+s} \langle (\phi_{\ell}^{\alpha})^2, |R|(X_N(u \wedge \tau_N)) \rangle_0 + \sum_{i=1}^n D \langle \partial_i \phi_{\ell}^{\alpha} \rangle^2, X_N(u \wedge \tau_N) \rangle_0 du \mid F_{N,t} \right\} \end{cases}$$

by Lemma A.2 in Kotelenetz [22]

$$\leq \tilde{K}s$$

for some $\tilde{K} < \infty$ since by Lemma 2.1 $H_{\alpha-1} \hookrightarrow H_0$ is Hilbert-Schmidt and $||| X_N(t \wedge \tau_N) ||| \leq K + 1 < \infty$.

Setting

$$Y_{N,\hat{t}}(s) := \tilde{K}s + \frac{1}{s} \{ \tau_N < \hat{t}+s \}$$

we obtain from Theorem 3.1

$$\lim_{s \rightarrow 0} \overline{\lim_{N \rightarrow \infty}} E Y_{N,\hat{t}}(s) = 0.$$

(iii) (i) and (ii) imply by Theorem 2.7 of Kurtz [29] the weak convergence of M_N to M . □

Theorem 3.2

Under the assumptions of Theorem 3.1 for all $\alpha > \frac{n}{2} + 1$

$$\int_0^{\cdot} U_N(\cdot, s) dM_N(s) \Rightarrow \int_0^{\cdot} U_{-\alpha}(\cdot, s) dM(s) \quad \text{on } D([0, \infty); H_{-\alpha}).$$

Proof

(i) Let d_p denote the Prohorov metric on $D([0, \infty); H_{-\alpha})$ (cf. Billingsley [4]). Let π_k be the projection of $H_{-\alpha}$ onto $L(\phi_{\ell} : \ell_i < k \text{ for all } i = 1, \dots, n)$ (the linear hull spanned by those ϕ_{ℓ} whose multiindices

$\ell = (\ell_1, \dots, \ell_n)$ satisfy $\ell_i < k$ for all $i = 1, \dots, n$. The corresponding projection from $H_{0,N}$ onto $L(\phi_{\ell,N}: \ell_i < k \text{ for all } i = 1, \dots, n)$ if $k < h^{-1}$ will be denoted by p_k^N . If $k \geq h^{-1}$ then we set $p_k^N H_{0,N} = H_{0,N}$. Set $\pi_k^\perp = I - \pi_k$ and $p_k^{\perp N} = I - p_k^N$, where I denotes the identity operator on the corresponding spaces.

(ii) Abbreviating the convolution integrals $\int_0^\cdot U_N(\cdot, s) dM_N(s)$ by $\int U_N dM_N$ etc.

we obtain

$$\begin{aligned}
 & d_p(\int U_N dM_N, \int U dM) \\
 & \leq d_p(\int U_N dM_N, \int U_N p_k^N dM_N) \\
 (*) \quad & + d_p(\int U_N p_k^N dM_N, \int U \pi_k dM) \\
 & + d_p(\int U \pi_k dM, \int U dM).
 \end{aligned}$$

(iii) By Lemma 2.3 and (2.15) for any $\hat{t} \geq 0$

$$\begin{aligned}
 & E \sup_{0 \leq t \leq \hat{t}} \left| \int_0^t U_N(t, s) p_k^{\perp N} dM_N \right|_{-\alpha}^2 \\
 & \leq c e^{4\beta \hat{t}} E |p_k^{\perp N} M_N(\hat{t})|_{-\alpha}^2 \\
 & \leq c e^{4\beta \hat{t}} \sum_{\ell_i \geq k} E < M_N(\hat{t}), \phi_{\ell N} >_0^2 \lambda_\ell^{-\alpha} \\
 & \leq c e^{4\beta \hat{t}} K(\hat{t}+1) \sum_{\ell_i \geq k} \lambda_\ell^{-\alpha+1}
 \end{aligned}$$

as in the proof of Lemma 3.3 for some finite constant K . Since $\sum \lambda_\ell^{-\alpha+1} < \infty$ by Lemma 2.1 the r.h.s. of the last inequality can be made arbitrarily small by choosing k large. Hence for given $\varepsilon > 0$ there is a $k_1(\varepsilon)$ s.t. for all $k \geq k_1(\varepsilon)$ and all N

$$d_p(\int U_N dM_N, \int U_N p_k^N dM_N) \leq \frac{\varepsilon}{3}$$

(cf. Kotelenez [24], [25]).

The third term in (*) can be estimated in the same way. The second term in (*) tends to zero for fixed k . Indeed, by partial integration

$$\int_0^t U_N(t,s) p_k^N dM_N(s) = p_k^N M_N(t) + \int_0^t U_N(t,s) [D\Delta_N + R'(X(s))] p_k^N M_N(s) ds \quad \text{and}$$

$$\int_0^t U(t,s) \pi_k dM(s) = \pi_k M(t) + \int_0^t U(t,s) (D\Delta + R'(X(s))) \pi_k M(s) ds. \quad \text{Hence, the}$$

Trotter-Kato theorem (Davies [9], Theorem 3.17, Kotelenez [24], Remark 4.1, and Kotelenez [22], Lemmas A.1, A.3) and the definition of p_k^N and π_k imply the conditions of Theorem 5.5 in Billingsley [4], Ch.I.

(iv) Since weak convergence on $D([0,\infty); \mathbb{H}_{-\alpha})$ and convergence w.r.t. the Prohorov metric d_p are equivalent (cf. Kurtz [29] and Billingsley [4], Appendix III, Th. 5) the proof is finished. \square

Fix

$$\alpha > \frac{n}{2} + 1.$$

Let Y_0 be an $\mathbb{H}_{-\alpha+1}$ -valued square integrable random variable on $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathcal{F}}_t, \tilde{\mathbb{P}})$ independent of $W(t)$ for all $t \geq 0$ and \underline{Y}_0 a square integrable $\mathbb{H}_{-\alpha+1}$ -valued random variable on $(\Omega, \mathcal{F}, \mathcal{F}_t, \mathbb{P})$ such $\underline{Y}_0 \stackrel{D}{=} Y_0$. Further, let C denote an arbitrary $\int U F_{-\alpha+1}^{1/2} dW$ - continuity set of $D([0,\infty); \mathbb{H}_{-\alpha})$ (cf. Billingsley [4]) and E an arbitrary element from $\sigma(Y_0)$. We make the following asymptotic independence assumption:

$$(3.5) \quad \begin{aligned} & Y_{N,0} := (vN)^{1/2} (X_{N,0} - X_0) \rightarrow Y_0 \text{ in probability on } \mathbb{H}_{-\alpha+1} \\ & P\{(\int U_N dM_N \in C) \cap E\} \rightarrow \tilde{P}\{(\int U F_{-\alpha+1}^{1/2} dW \in C) \tilde{P}(E) \} \end{aligned}$$

(cf. Billingsley [4], Ch.I, Th. 2.1 - The second condition in (3.5) is, e.g., satisfied if M_N is independent of $\sigma(Y_0)$ for all (large) N .

Let δ denote the Fréchet derivative, $B([0, \hat{t}] \times H_{-\alpha})$ the real valued measurable functions g with domain $[0, \hat{t}] \times H_{-\alpha}$ s.t. $\frac{\partial g}{\partial t}$, δg , $\delta^2 g$, and $D\Delta_{-\alpha} \delta g$ exist, are continuous in x and t , and uniformly bounded in norm on $[0, T] \times H_{-\alpha}$. $Q^{1/2}$ is the square root of the covariance operator of $W(t)$ on $H_{-\alpha+1}$ and $F_{-\alpha+1}^{1/2*}(X(t))$ is the dual operator of $F_{-\alpha+1}^{1/2}(X(t))$ (after identifying the duals of $H_{-\alpha}$ and $H_{-\alpha+1}$ with $H_{-\alpha}$ and $H_{-\alpha+1}$, respectively. Finally, "Tr" denotes "trace".

Now we can state our final result under the assumptions of the LLN.

Theorem 3.3 (CLT)

Assume (2.9) and (3.5) for fixed $\alpha > \frac{n}{2} + 1$ in addition to

$$(I) \quad v = N^p \text{ where } p > \frac{2\gamma+1+\frac{2}{n}}{1-2\gamma} \text{ and } \gamma \in [\frac{1}{4}, \frac{1}{2}) \text{ arbitrary and fixed ;}$$

$$(II) \quad E |(vN)^{\gamma} (X_N(0) - X(0))|_0^2 \rightarrow 0, \text{ as } N \rightarrow \infty.$$

Then for $Y_N := (vN)^{1/2} (X_N - X)$

$$(i) \quad Y_N \Rightarrow Y \text{ on } D([0, \infty); H_{-\alpha})$$

where

$$(3.6) \quad Y(t) = U_{-\alpha}(t, 0) Y_0 + \int_0^t U_{-\alpha}(t, s) F_{-\alpha+1}^{1/2}(X(s)) dW(s)$$

is the mild solution of the stochastic partial differential equation

$$(3.7) \quad \begin{aligned} dY(t) &= [D\Delta_{-\alpha} + R'(X(t))]Y(t)dt + F_{-\alpha+1}^{1/2}(X(t))dW(t) \\ Y(0) &= Y_0 \end{aligned}$$

(ii)

$$(3.8) \quad Y \in \begin{cases} C^\mu([0, \hat{t}]; H_{-\alpha}) \text{ a.s. for all } \mu < \frac{1}{2}, \text{ all } \hat{t} > 0 \\ C([0, \hat{t}]; H_{-\alpha+1}) \text{ a.s. for all } \hat{t} > 0. \end{cases}$$

and $Y(t)$, $t > 0$, does not define a σ -additive measure on $H_{-\alpha}$ for $\gamma \leq \frac{n}{2}$, i.e., the second relation in (3.8) is the maximal regularity of Y on the Hilbert scale (2.3).

(iii) Y is a Markov process, and its weak generator is given by

$$(3.9) \quad \begin{aligned} A(t)g(t, \varphi') &= \frac{\partial}{\partial t} g(t, \varphi') + \langle [D\Delta_{-\alpha} + R'(X(t))] \delta g(t, \varphi'), \varphi' \rangle_{-\alpha} \\ &+ \frac{1}{2} \text{Tr} \{ Q^{1/2} F_{-\alpha+1}^{1/2*}(X(t)) \delta^2 g(t, \varphi') F_{-\alpha+1}^{1/2}(X(t)) Q^{1/2} \}, \end{aligned}$$

where $g \in B([0, \hat{t}] \times H_{-\alpha})$.

Proof

(i) The norm of the normalized last term in (2.13) can be estimated as follows:

$$\begin{aligned} & \left| \int_0^t U_N(t, s) (X_N(s) - X(s))^2 (vN)^{1/2} \tilde{R}(X_N(s), X(s)) \right|_{-\alpha} \\ & \leq \int_0^t e^{\beta(t-s)} \left| \left| (vN)^{1/4} (X_N(s) - X(s)) \right| \right|^2 \cdot \left| \tilde{R}(X_N(s), X(s)) \right| ds \rightarrow 0, \end{aligned}$$

as $N \rightarrow \infty$ uniformly on compact intervals in probability by Theorem 3.1 and the stochastic boundedness of $\left| \tilde{R}(X_N(s), X(s)) \right|$ (cf. the proof of Theorem 3.1). Therefore, the weak convergence of Y_N to Y follows from Theorem 3.2 and our assumptions as in Kotelenetz [25], [26].

(ii) The Hölder continuity follows from DaPrato, Iannelli and Tubaro [8] and Kotelenetz [23].

The spatial regularity follows from the estimate

$$\begin{aligned} \left| \int_s^t U_{-\alpha}(t,u) F_{-\alpha+1}^{1/2}(X(u)) dW(u) \right|_{-\alpha+1} &\leq \left| \int_s^t T_{-\alpha}(t-u) F_{-\alpha+1}^{1/2}(X(u)) dW(u) \right|_{-\alpha+1} \\ &+ \int_s^t e^{\beta(t-u)} \left| \int_s^u T_{-\alpha}(u-v) F_{-\alpha+1}^{1/2}(X(v)) dW(v) \right|_{-\alpha+1} du \end{aligned}$$

and the spatial regularity of $\int_0^\cdot T_{-\alpha}(\cdot-s) F_{-\alpha+1}^{1/2}(X(s)) dW(s)$, as proved in Kotelenetz [27] (generalizing a result of Dawson [10] - cf. also Kotelenetz [25]). That the spatial regularity in (3.8) is maximal follows from the Gaussianity of $\int_0^t U_{-\alpha}(t,s) F_{-\alpha+1}^{1/2}(X(s)) dW(s)$ - cf. for details Kotelenetz [25].

(iii) The Markov property follows from Arnold, Curtain and Kotelenetz [2], (3.9) follows from Curtain [6].

Remark 3.1

I. The final result can be expressed by

$$(3.10) \quad X_N = X + \frac{1}{\sqrt{vN}} Y + o\left(\frac{1}{\sqrt{vN}}\right),$$

where X_N is the local stochastic, i.e., mesoscopic description, assuming $X_N(0)$ being near to homogeneity, X is the deterministic homogeneous state solution of (2.1), and Y is the mild solution of (3.7) which is a generalized Gauss-Markov process if Y_0 is Gaussian, and $o(\frac{1}{\sqrt{vN}})$ is the error term.

II. Let us now assume that we do not start in (2.1) with a constant but with some other positive bounded and possibly smooth function $X_0(q)$.

Then the difference $X_N(t) - X(t)$ satisfies

$$(3.11) \quad X_N(t) - X(t) = F_N(t) + G_N(t),$$

where $F_N(t)$ is the r.h.s. in (2.13) and $G_N(t) = \int_0^t U_N(t,s) (D\Delta_N - D\Delta) X(s) ds$.

Of course, (3.11) will also tend to zero under the assumptions of Theorem 3.1. However, in view of (2.18) we must normalize (3.11) by multiplying both sides by $(vN)^{1/2}$ (modulus a constant) in order to obtain a Gaussian correction term. On the other hand, the convergence of $(vN)^{1/2} G_N(t)$ to zero with $v = N^p$ and $p > 1$ does not hold (in general) in function norms and for $p < 2$ Arnold and Theodosopulu [3] have shown in the one-dimensional case that the variance of $Z_N(t)$ (the martingale part of $X_N(t)$) tends to ∞ in L_2 -norm. This problem and related questions will be investigated in a forthcoming paper.

References

- [1] ARNOLD, L. (1981) *Mathematical Models of Chemical Reactions*, in M. Hazewinkel, J. Willems (eds.) "Stochastic Systems", Dordrecht.
- [2] ARNOLD, L., CURTAIN, R.F., KOTELNEZ, P. (1980) *Nonlinear Evolution Equations in Hilbert Space*, Forschungsschwerpunkt Dynamische Systeme, Universität Bremen, Report Nr. 17.
- [3] ARNOLD, L., THEODOSOPULU, M. (1980) *Deterministic Limit of the Stochastic Model of Chemical Reactions with Diffusion*, Adv. Appl. Prob. 12, 367-379.
- [4] BILLINGSLEY, P. (1968) *Convergence of Probability Measures*, John Wiley & Sons, New York.
- [5] CODDINGTON, E.A. AND LEVINSON, N. (1955) *Theory of Ordinary Differential Equations*, McGraw-Hill, New York.
- [6] CURTAIN, R.F. (1981) *Markov Processes Generated by Linear Stochastic Evolution Equations*, Stochastic 5, 135-165.
- [7] CURTAIN, R.F., PRITCHARD, A.J. (1978) *Infinite Dimensional Linear Systems Theory*, LN in Control and Information Sciences 8, Springer-Verlag, Berlin-New York.
- [8] DAPRATO, G., IANNELLI, M., TUBARO, L. (1982) *On the Path Regularity of a Stochastic Process in a Hilbert Space Defined by the Itô Integral*, Stochastics 6, 315-322.
- [9] DAVIES, E.B. (1980) *One-Parameter Semigroups*, Academic Press, London-New York.
- [10] DAWSON, D.A. (1972) *Stochastic Evolution Equations*, Math. Biosciences 15, 287-316.
- [11] DAWSON, D.A. (1975) *Stochastic Evolution Equations and Related Measure Processes*, J. multivariate Analysis 5, 1-52.
- [12] GARDINER, C.W., McNEIL, K.J., WALLS, D.F., MATHESON, I.S. (1976) *Correlations in Stochastic Theories of Chemical Reactions*, Journal of Statistical Physics, Vol. 14 No.4, 307-331.
- [13] GIHMAN, I.I., SKOROHOD, A.V. (1974) *The Theory of Stochastic Processes*, Vol. 2, Springer-Verlag, Berlin-New York.
- [14] GOROSTIZA, L.G. (1983) *High Density Limit Theorems for Infinite Systems of Unscaled Branching Brownian Motions*, Ann. of Probab. 11(2), 374-392.

- [15] HAKEN, H. (1983) *Advanced Synergetics*, Springer-Verlag, Berlin-New York.
- [16] HOLLEY, R., STROOCK, D.W. (1978) *Generalized Ornstein-Uhlenbeck Processes and Infinite Particle Branching Brownian Motions*, Publ. RIMS, Kyoto, Univ. 14, 741-788.
- [17] ITÔ, K. (1980) *Continuous Additive S^1 -Processes*, in B. Grigelionis (ed.) "Stochastic Differential Systems" Springer-Verlag, Berlin-New York.
- [18] ITÔ, K. (1983) *Distribution-Valued Processes Arising from Independent Brownian Motions*, Math. Z. 182, 17-33.
- [19] KALLIANPUR, G. AND WOLPERT, R. (1984) *Infinite Dimensional Differential Equation Models for Spatially Distributed Neurons*, Appl. Math. Optim. 12, 125-172.
- [20] VAN KAMPEN, N.G. (1983) *Stochastic Processes in Physics and Chemistry*, North-Holland, Amsterdam-New York.
- [21] KOTELENEZ, P. (1982) *A Submartingale Type Inequality with Applications to Stochastic Evolution Equations*, Stochastics 8, 139-151.
- [22] KOTELENEZ, P. (1982) Ph.D. Thesis, Bremen.
- [23] KOTELENEZ, P. (1984) *Continuity Properties of Hilbert Space Valued Martingales*, Stochastic Processes Appl. 17 (1), 115-125.
- [24] KOTELENEZ, P. (1984) *A Stopped Doob Inequality for Stochastic Convolution Integrals and Stochastic Evolution Equations*, Stochastic Analysis and Applications, 2(3), 245-265.
- [25] KOTELENEZ, P. (1984) *Law of Large Numbers and Central Limit Theorem for Linear Chemical Reactions with Diffusion*, to appear in Ann. of Probab.
- [26] KOTELENEZ, P. (1984) *Linear Parabolic Differential Equations as Limits of Space-Time Jump Markov Processes*, to appear in J. Math. Anal. Appl.
- [27] KOTELENEZ, P. (1985) *On the Semigroup Approach to Stochastic Evolution Equations*, in L. Arnold, P. Kotelenez (eds.) "Stochastic Space-Time Models and Limit Theorems", Reidel Publ., Dordrecht, Holland.
- [28] KURTZ, T. (1971) *Limit Theorems for Sequences of Jump Markov Processes Approximating Ordinary Differential Processes*, J. Appl. Prob. 8, 344-356.
- [29] KURTZ, T. (1981) *Approximation of Population Processes*, CBMS-NSF Regional Conference Series in Applied Mathematics, 36, SIAM.

- [30] MARTIN-LÖF, A. (1976) *Limit Theorems for the Motions of a Poisson System of Independent Markovian Particles with High Density*, Z. Wahrscheinlichkeitstheorie verw. Gebiete 34, 205-223.
- [31] NICOLIS, G. AND PRIGOGINE, I. (1977) *Self-Organization in Non-equilibrium Systems*, John Wiley & Sons, New York-London.
- [32] SCHWARTZ, L. (1954) *Sur l'impossibilité de la multiplication des distributions*, *C.R.Acad. Sci.*, Paris 239, 847-848.
- [33] SMOLLER, J. (1983) *Shock Waves and Reaction-Diffusion Equations*, Springer-Verlag, New York-Berlin.
- [34] TANABE, H. (1979) *Equations of Evolution*, Pitman, London-San Francisco-Melbourne.
- [35] USTUNEL, A.S. (1984) *Additive Processes on Nuclear Spaces*, *Ann. of Probab.* 12, Nr. 3, 858-868.
- [36] WALSH, J. (1981) *A Stochastic Model of Neural Response*, *Adv. Appl. Prob.* 13, 231-281.
- [37] WALSH, J. (1985) *An Introduction to Stochastic Partial Differential Equations*, to appear
- [38] YOSIDA, K. (1968) *Functional Analysis*, Springer-Verlag, Berlin-New York.

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